In [1]: pip install gdown pandas matplotlib numpy

```
Requirement already satisfied: gdown in /Users/sachinladdha/anaconda3/lib/python3.1
2/site-packages (5.2.0)
Requirement already satisfied: pandas in /Users/sachinladdha/anaconda3/lib/python3.1
2/site-packages (2.2.2)
Requirement already satisfied: matplotlib in /Users/sachinladdha/anaconda3/lib/pytho
n3.12/site-packages (3.9.2)
Requirement already satisfied: numpy in /Users/sachinladdha/anaconda3/lib/python3.1
2/site-packages (1.26.4)
Requirement already satisfied: beautifulsoup4 in /Users/sachinladdha/anaconda3/lib/p
ython3.12/site-packages (from gdown) (4.12.3)
Requirement already satisfied: filelock in /Users/sachinladdha/anaconda3/lib/python
3.12/site-packages (from gdown) (3.13.1)
Requirement already satisfied: requests[socks] in /Users/sachinladdha/anaconda3/lib/
python3.12/site-packages (from gdown) (2.32.3)
Requirement already satisfied: tqdm in /Users/sachinladdha/anaconda3/lib/python3.12/
site-packages (from gdown) (4.66.5)
Requirement already satisfied: python-dateutil>=2.8.2 in /Users/sachinladdha/anacond
a3/lib/python3.12/site-packages (from pandas) (2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in /Users/sachinladdha/anaconda3/lib/pyt
hon3.12/site-packages (from pandas) (2024.1)
Requirement already satisfied: tzdata>=2022.7 in /Users/sachinladdha/anaconda3/lib/p
ython3.12/site-packages (from pandas) (2023.3)
Requirement already satisfied: contourpy>=1.0.1 in /Users/sachinladdha/anaconda3/li
b/python3.12/site-packages (from matplotlib) (1.2.0)
Requirement already satisfied: cycler>=0.10 in /Users/sachinladdha/anaconda3/lib/pyt
hon3.12/site-packages (from matplotlib) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /Users/sachinladdha/anaconda3/li
b/python3.12/site-packages (from matplotlib) (4.51.0)
Requirement already satisfied: kiwisolver>=1.3.1 in /Users/sachinladdha/anaconda3/li
b/python3.12/site-packages (from matplotlib) (1.4.4)
Requirement already satisfied: packaging>=20.0 in /Users/sachinladdha/anaconda3/lib/
python3.12/site-packages (from matplotlib) (24.1)
Requirement already satisfied: pillow>=8 in /Users/sachinladdha/anaconda3/lib/python
3.12/site-packages (from matplotlib) (10.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in /Users/sachinladdha/anaconda3/li
b/python3.12/site-packages (from matplotlib) (3.1.2)
Requirement already satisfied: six>=1.5 in /Users/sachinladdha/anaconda3/lib/python
3.12/site-packages (from python-dateutil>=2.8.2->pandas) (1.16.0)
Requirement already satisfied: soupsieve>1.2 in /Users/sachinladdha/anaconda3/lib/py
thon3.12/site-packages (from beautifulsoup4->gdown) (2.5)
Requirement already satisfied: charset-normalizer<4,>=2 in /Users/sachinladdha/anaco
nda3/lib/python3.12/site-packages (from requests[socks]->gdown) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /Users/sachinladdha/anaconda3/lib/pyt
hon3.12/site-packages (from requests[socks]->gdown) (3.7)
Requirement already satisfied: urllib3<3,>=1.21.1 in /Users/sachinladdha/anaconda3/l
ib/python3.12/site-packages (from requests[socks]->gdown) (2.2.3)
Requirement already satisfied: certifi>=2017.4.17 in /Users/sachinladdha/anaconda3/l
ib/python3.12/site-packages (from requests[socks]->gdown) (2024.8.30)
Requirement already satisfied: PySocks!=1.5.7,>=1.5.6 in /Users/sachinladdha/anacond
a3/lib/python3.12/site-packages (from requests[socks]->gdown) (1.7.1)
Note: you may need to restart the kernel to use updated packages.
```

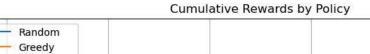
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

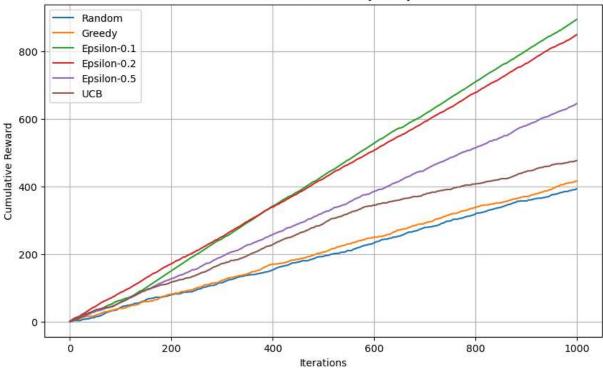
```
from collections import defaultdict
        import gdown
In [3]: # Download the dataset from Google Drive
        file_id = "1gfobhq1VCw80o52JCiYpEBGhG5k7cWBr"
        url = f"https://drive.google.com/uc?id={file id}&export=download"
        dataset_path = "movie_ratings.csv"
        gdown.download(url, dataset path, quiet=False)
       Downloading...
       From: https://drive.google.com/uc?id=1gfobhqlVCw8Oo52JCiYpEBGhG5k7cWBr&export=downlo
       To: /Users/sachinladdha/study/BITS/Deep Reinforcement Learning/Assignment-1/movie ra
       tings.csv
       100%
                                            2.48M/2.48M [00:00<00:00, 4.11MB/s]
Out[3]: 'movie_ratings.csv'
In [4]: # Load and preprocess the dataset (Step 1)
        df = pd.read_csv(dataset_path)
In [5]: # Binarize rewards: Rating >= 4 -> Reward = 1, else Reward = 0
        df['Reward'] = (df['rating'] >= 4).astype(int)
In [6]: # Define the environment and compute rewards (Step 2)
        movies = df['movieId'].unique()
        movie rewards = {movie: [] for movie in movies}
        for movie in movies:
            movie rewards[movie] = df[df['movieId'] == movie]['Reward'].tolist()
        def get_reward(movie_id):
            rewards = movie_rewards[movie_id]
            return rewards[np.random.randint(len(rewards))]
In [7]: # Simulation setup
        num_iterations = 1000
        cumulative rewards = defaultdict(list)
In [8]: # Random Policy (Step 3)
        def random_policy():
            total\_reward = 0
            for _ in range(num_iterations):
                movie = np.random.choice(movies)
                reward = get_reward(movie)
                total_reward += reward
                cumulative_rewards['Random'].append(total_reward)
        random_policy()
In [9]: # Greedy Policy (Step 4)
        def greedy_policy():
            total reward = 0
            estimated_rewards = {movie: 0 for movie in movies}
            counts = {movie: 0 for movie in movies}
```

```
if 0 in counts.values(): # If an arm hasn't been tried yet, try it
                     movie = np.random.choice([m for m in movies if counts[m] == 0])
                 else:
                     movie = max(movies, key=lambda m: estimated_rewards[m])
                 reward = get reward(movie)
                 counts[movie] += 1
                 estimated_rewards[movie] += (reward - estimated_rewards[movie]) / counts[mo
                 total reward += reward
                 cumulative_rewards['Greedy'].append(total_reward)
         greedy_policy()
In [10]: # Epsilon-Greedy Policy (Step 5)
         def epsilon_greedy_policy(epsilon):
             total_reward = 0
             estimated rewards = {movie: 0 for movie in movies}
             counts = {movie: 0 for movie in movies}
             for in range(num iterations):
                 if np.random.rand() < epsilon: # Explore</pre>
                     movie = np.random.choice(movies)
                 else: # Exploit
                     movie = max(movies, key=lambda m: estimated rewards[m])
                 reward = get reward(movie)
                 counts[movie] += 1
                 estimated_rewards[movie] += (reward - estimated_rewards[movie]) / counts[mo
                 total reward += reward
                 cumulative_rewards[f'Epsilon-{epsilon}'].append(total_reward)
         for epsilon in [0.1, 0.2, 0.5]:
             epsilon_greedy_policy(epsilon)
In [11]: # UCB Algorithm (Step 6)
         def ucb_policy():
             total reward = 0
             estimated_rewards = {movie: 0 for movie in movies}
             counts = {movie: 0 for movie in movies}
             for t in range(1, num_iterations + 1):
                 movie = max(movies, key=lambda m: estimated rewards[m] + np.sqrt(2 * np.log
                 reward = get_reward(movie)
                 counts[movie] += 1
                 estimated rewards[movie] += (reward - estimated rewards[movie]) / counts[mo
                 total reward += reward
                 cumulative_rewards['UCB'].append(total_reward)
         ucb policy()
In [12]: # Plot Cumulative Rewards (Step 7)
         plt.figure(figsize=(10, 6))
         for policy, rewards in cumulative_rewards.items():
             plt.plot(rewards, label=policy)
         plt.xlabel('Iterations')
```

for \_ in range(num\_iterations):

```
plt.ylabel('Cumulative Reward')
plt.title('Cumulative Rewards by Policy')
plt.legend()
plt.grid()
plt.show()
```





```
In [13]: # Conclusion (Step 8)
         print("\nConclusion:")
         print("The policy with the highest cumulative reward is:", max(cumulative_rewards,
```

## Conclusion:

The policy with the highest cumulative reward is: Epsilon-0.1

```
In [ ]:
```